

1 Directional Derivatives

Definition. Given a function $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ (i.e. $z = f(x, y)$), a point (x_0, y_0) and a unit vector $\vec{u} = \langle u_1, u_2 \rangle$, define the directional derivative of f in the direction \vec{u} at (x_0, y_0) , denoted $D_{\vec{u}}f(x_0, y_0)$, to be

$$D_{\vec{u}}f(x_0, y_0) = \lim_{t \rightarrow 0} \frac{f(x_0 + tu_1, y_0 + tu_2) - f(x_0, y_0)}{t}$$

(if this limit exists).

A directional derivative measures the rate of change of f in the direction indicated by \vec{u} at the point (x_0, y_0) . In other words, look at the point (x_0, y_0) in the xy -plane and draw the vector \vec{u} starting at (x_0, y_0) . Look at the whole line in the xy -plane along this vector. Imagine you sliced through the graph of f straight down, through this line. You get a curve; the directional derivative is the slope of the tangent line to this curve when $x = x_0, y = y_0$.

We have seen two examples of directional derivatives, namely partial derivatives:

- Suppose $\vec{u} = \langle 1, 0 \rangle$. Then

$$D_{\vec{u}}f(x_0, y_0) = \lim_{t \rightarrow 0} \frac{f(x_0 + t, y_0) - f(x_0, y_0)}{t} = f_x(x_0, y_0).$$

- Suppose $\vec{u} = \langle 0, 1 \rangle$. Then

$$D_{\vec{u}}f(x_0, y_0) = \lim_{t \rightarrow 0} \frac{f(x_0, y_0 + t) - f(x_0, y_0)}{t} = f_y(x_0, y_0).$$

The following theorem shows you how to calculate directional derivatives:

Theorem. Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$. If (x_0, y_0) and $\vec{u} = \langle u_1, u_2 \rangle$ are such that $D_{\vec{u}}f(x_0, y_0)$ exists, then

$$D_{\vec{u}}f(x_0, y_0) = \langle f_x(x_0, y_0), f_y(x_0, y_0) \rangle \cdot \vec{u}.$$

Proof: Define the function $t \mapsto (x, y)$ by

$$x(t) = x_0 + tu_1, y(t) = y_0 + tu_2.$$

Now we have (from the definition of directional derivative)

$$D_{\vec{u}}f(x_0, y_0) = \lim_{t \rightarrow 0} \frac{f(x(t), y(t)) - f(x(0), y(0))}{t}.$$

Notice that this is the derivative $\frac{df}{dt}$ of the function $f(x(t), y(t))$ at $t = 0$. So by a Chain Rule from the previous section, we have

$$D_{\vec{u}} f(x_0, y_0) = \left. \frac{df}{dt} \right|_{t=0} = \left(\frac{\partial f}{\partial x} \frac{dx}{dt} + \frac{\partial f}{\partial y} \frac{dy}{dt} \right) \Big|_{t=0}.$$

However, we can explicitly calculate $\frac{dx}{dt}$ and $\frac{dy}{dt}$ since we know the formulas for $x(t)$ and $y(t)$. Also, using the formulas for $x(t)$ and $y(t)$ we know at $t = 0$ that $x(0) = x_0$ and $y(0) = y_0$. We have:

$$D_{\vec{u}} f(x_0, y_0) = \left(\frac{\partial f}{\partial x}(x_0, y_0)u_1 + \frac{\partial f}{\partial y}(x_0, y_0)u_2 \right) = \langle f_x(x_0, y_0), f_y(x_0, y_0) \rangle \cdot \langle u_1, u_2 \rangle$$

so the theorem is proved. ■

Example: Let $f(x, y) = 4x^2y^3 - 2xy$. Find the directional derivative of f in the direction $\langle 1, -1 \rangle$ at the point $(2, 1)$.

Solution: First find the partial derivatives:

$$f_x(x, y) = 8xy^3 - 2y \quad f_y(x, y) = 12x^2y^2 - 2x$$

Next, you need a unit vector \vec{u} in the direction of $\langle 1, -1 \rangle$. To do this, take $\langle 1, -1 \rangle$ and divide by its length (which is $\sqrt{2}$). You get

$$\vec{u} = \frac{\langle 1, -1 \rangle}{\sqrt{2}} = \left\langle \frac{1}{\sqrt{2}}, \frac{-1}{\sqrt{2}} \right\rangle.$$

By the theorem,

$$\begin{aligned} D_{\vec{u}} f(2, 1) &= \langle f_x(2, 1), f_y(2, 1) \rangle \cdot \left\langle \frac{1}{\sqrt{2}}, \frac{-1}{\sqrt{2}} \right\rangle \\ &= \langle 8(2)(1^3) - 2(1), 12(2^2)(1^2) - 2(1) \rangle \cdot \left\langle \frac{1}{\sqrt{2}}, \frac{-1}{\sqrt{2}} \right\rangle \\ &= \langle 14, 46 \rangle \cdot \left\langle \frac{1}{\sqrt{2}}, \frac{-1}{\sqrt{2}} \right\rangle \\ &= \frac{14 - 46}{\sqrt{2}} = -16\sqrt{2}. \end{aligned}$$

You can also define directional derivatives for functions of three (or more) variables in a similar way:

Definition. Given a function $f : \mathbb{R}^3 \rightarrow \mathbb{R}$ (i.e. the function looks like $f(x, y, z)$), a point (x_0, y_0, z_0) and a vector $\vec{u} = \langle u_1, u_2, u_3 \rangle$, define the directional derivative of f in the direction \vec{u} at (x_0, y_0, z_0) , denoted $D_{\vec{u}} f(x_0, y_0, z_0)$, to be

$$D_{\vec{u}} f(x_0, y_0, z_0) = \lim_{t \rightarrow 0} \frac{f(x_0 + tu_1, y_0 + tu_2, z_0 + tu_3) - f(x_0, y_0, z_0)}{t}$$

(if this limit exists).

The same theorem works:

Theorem. Let $f : \mathbb{R}^3 \rightarrow \mathbb{R}$. If (x_0, y_0, z_0) and $\vec{u} = \langle u_1, u_2, u_3 \rangle$ are such that $D_{\vec{u}}f(x_0, y_0, z_0)$ exists, then

$$D_{\vec{u}}f(x_0, y_0, z_0) = \langle f_x(x_0, y_0, z_0), f_y(x_0, y_0, z_0), f_z(x_0, y_0, z_0) \rangle \cdot \vec{u}.$$

2 Gradients

Notice that in the two theorems above we saw the following expressions appear:

- $\langle f_x(x, y), f_y(x, y) \rangle$ (for functions of two variables)
- $\langle f_x(x, y, z), f_y(x, y, z), f_z(x, y, z) \rangle$ (for functions of three variables)

Vectors of this type are important for many reasons, so we give them a name:

Definition. Given a function $f : \mathbb{R}^2 \rightarrow \mathbb{R}$, define the gradient of f , denoted ∇f or $\text{grad } f$, to be

$$\nabla f = \langle f_x, f_y \rangle.$$

The gradient is a vector-valued function of x and y . In particular $\nabla f : \mathbb{R}^2 \rightarrow \mathbb{R}^2$.

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$$\nabla f = \langle f_x, f_y, f_z \rangle.$$

The gradient is a vector-valued function of x , y and z . In particular $\nabla f : \mathbb{R}^3 \rightarrow \mathbb{R}^3$.

Example: Find the gradient of $f(x, y, z) = e^{x \sin y \cos z}$.

Solution:

$$\begin{aligned} \nabla f &= \langle f_x, f_y, f_z \rangle \\ &= \langle \sin y \cos z e^{x \sin y \cos z}, x \cos y \cos z e^{x \sin y \cos z}, -x \sin y \sin z e^{x \sin y \cos z} \rangle. \end{aligned}$$

The two theorems in the previous section can be restated as one result in terms of gradients:

Theorem. Let f be a function of several variables (2, 3 or any number of variables). Then the directional derivative of f in the direction \vec{u} is given by

$$D_{\vec{u}}f = \nabla f \cdot \vec{u}.$$

Here are some other reasons why the gradient of a function is important. I give two theorems below for functions of two variables; the theorems also hold for functions of three (or more) variables.

Theorem. Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ be a function of two variables (i.e. $z = f(x, y)$). Fix a value of z , say $z = z_0$ and consider the level curve for f at height z_0 . This level curve has equation $f(x, y) = z_0$. For any point (x_0, y_0) on this level curve, ∇f is normal to the level curve at (x_0, y_0) .

Proof: Let \vec{u} be a unit vector tangent to the level curve $f(x, y) = z_0$ at (x_0, y_0) . Then $D_{\vec{u}}f(x_0, y_0) = 0$ since the height of the function $f(x, y)$ does not change in the direction of the level curve. By the theorem in the previous section,

$$0 = D_{\vec{u}}f(x_0, y_0) = \nabla f(x_0, y_0) \cdot \vec{u}.$$

Therefore ∇f is orthogonal to the tangent vector \vec{u} at (x_0, y_0) , so it is normal to the level curve. This proves the theorem. ■

Theorem. Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ be a function of two variables (i.e. $z = f(x, y)$) and let (x_0, y_0) be some point in the domain of f . Then:

- The direction in which the value of f increases most rapidly from the point $(x_0, y_0, f(x_0, y_0))$ is ∇f .
- The direction in which the value of f decreases most rapidly from the point $(x_0, y_0, f(x_0, y_0))$ is $-\nabla f$.

Proof: Notice that the direction \vec{u} which gives the largest rate of increase in the value f is the direction \vec{u} which maximizes the directional derivative $D_{\vec{u}}f(x_0, y_0)$. But

$$\begin{aligned} D_{\vec{u}}f(x_0, y_0) &= \nabla f(x_0, y_0) \cdot \vec{u} \\ &= \|\nabla f(x_0, y_0)\| \|\vec{u}\| \cos \theta \text{ where } \theta \text{ is the angle between } \vec{u} \text{ and } \nabla f \\ &= \|\nabla f(x_0, y_0)\| \cos \theta \text{ since } \vec{u} \text{ is a unit vector} \end{aligned}$$

To maximize this, we choose $\theta = 0$ so that $\cos \theta = 1$. This is tantamount to choosing \vec{u} in the same direction of $\nabla f(x_0, y_0)$. (To minimize this choose $\theta = \pi$ so $\cos \theta = -1$, i.e. choose \vec{u} in the opposite direction of $\nabla f(x_0, y_0)$.) ■

Example: Let $f(x, y) = \frac{x}{x^2+y^2}$. Find a normal vector to the level curve $f(x, y) = .5$ at the point $(1, 1)$.

Solution: This is asking for $\nabla f(1, 1)$; the answer is $\langle f_x(1, 1), f_y(1, 1) \rangle = \langle 0, -2 \rangle$.

Example: The temperature of a point (x, y) on a metal plate is given by $T(x, y) = 100 - 2x^2 + 3y^2$. If a heat-seeking particle is dropped on the plate at the point $(2, -4)$, find the direction the particle moves in (when it first starts to move).

Solution: The particle will move in the direction that the temperature increases most rapidly, i.e. in the direction of the gradient. This question is asking for $\nabla f(2, -4)$ which is $\langle f_x(2, -4), f_y(2, -4) \rangle = \langle -8, -24 \rangle$.